

## A: Datasheet

Algorithm: s1\_000

Developer: Samsung S1 Corp

Submission Date: 2021\_06\_03

Template size: 4096 bytes

Template time (2.5 percentile): 864 msec

Template time (median): 865 msec

Template time (97.5 percentile): 871 msec

Investigation:

Frontal mugshot ranking 49 (out of 279) -- FNIR(1600000, 0, 1) = 0.0024 vs. lowest 0.0009 from sensetime\_005

Mugshot webcam ranking 65 (out of 241) -- FNIR(1600000, 0, 1) = 0.0166 vs. lowest 0.0062 from sensetime\_005

Mugshot profile ranking 30 (out of 210) -- FNIR(1600000, 0, 1) = 0.2581 vs. lowest 0.0587 from xforwardai\_002

Immigration visa-border ranking 45 (out of 168) -- FNIR(1600000, 0, 1) = 0.0054 vs. lowest 0.0013 from visionlabs\_010

Immigration visa-kiosk ranking 21 (out of 165) -- FNIR(1600000, 0, 1) = 0.0896 vs. lowest 0.0568 from cloudwalk\_hr\_000

Identification:

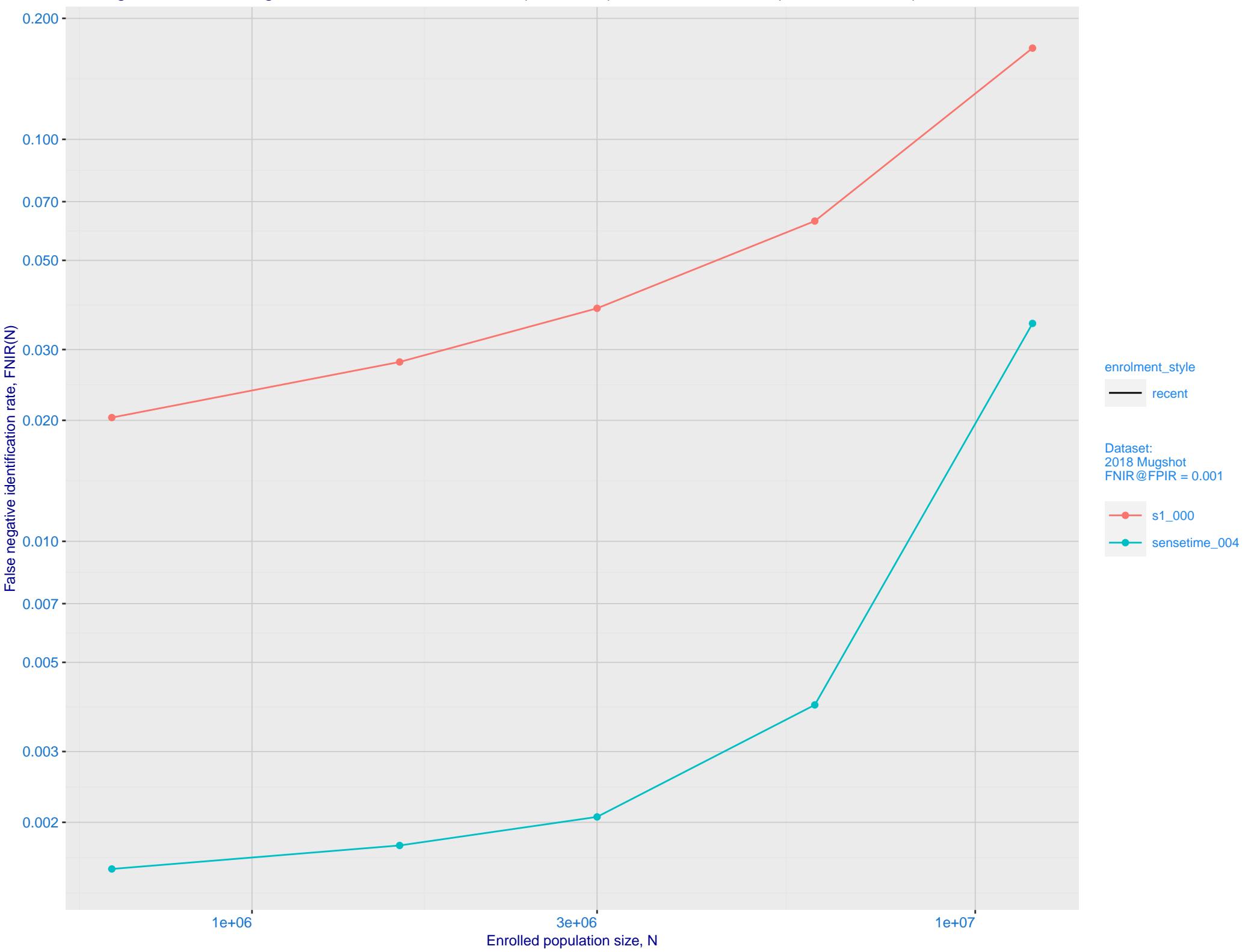
Frontal mugshot ranking 57 (out of 279) -- FNIR(1600000, T, L+1) = 0.0279, FPIR=0.001000 vs. lowest 0.0018 from sensetime\_004

Mugshot profile ranking 201 (out of 209) -- FNIR(1600000, T, L+1) = 1.0000, FPIR=0.001000 vs. lowest 0.1331 from cloudwalk\_hr\_000

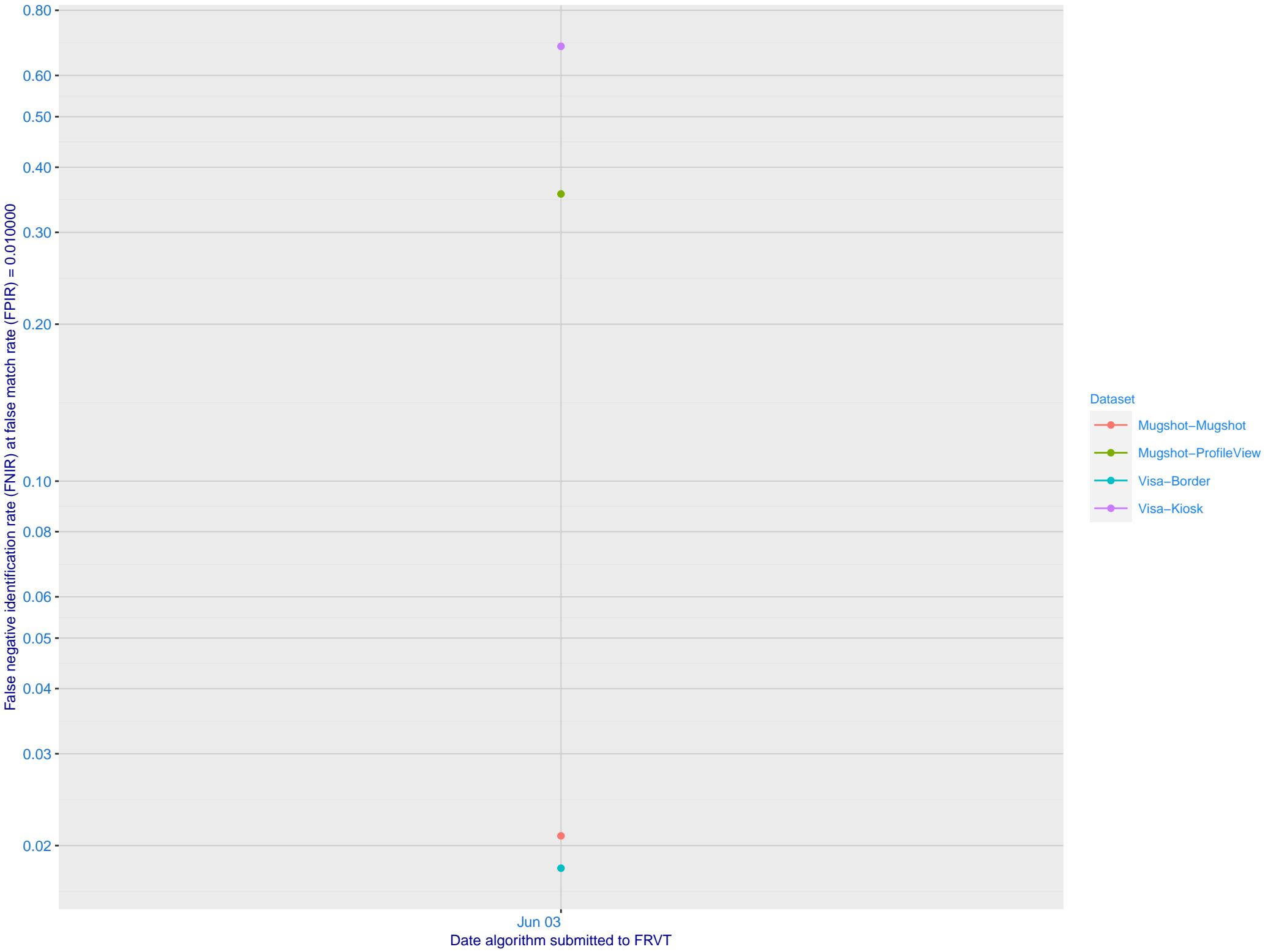
Immigration visa-border ranking 51 (out of 167) -- FNIR(1600000, T, L+1) = 0.0471, FPIR=0.001000 vs. lowest 0.0047 from idemia\_008

Immigration visa-kiosk ranking 154 (out of 162) -- FNIR(1600000, T, L+1) = 1.0000, FPIR=0.001000 vs. lowest 0.0996 from cloudwalk\_hr\_000

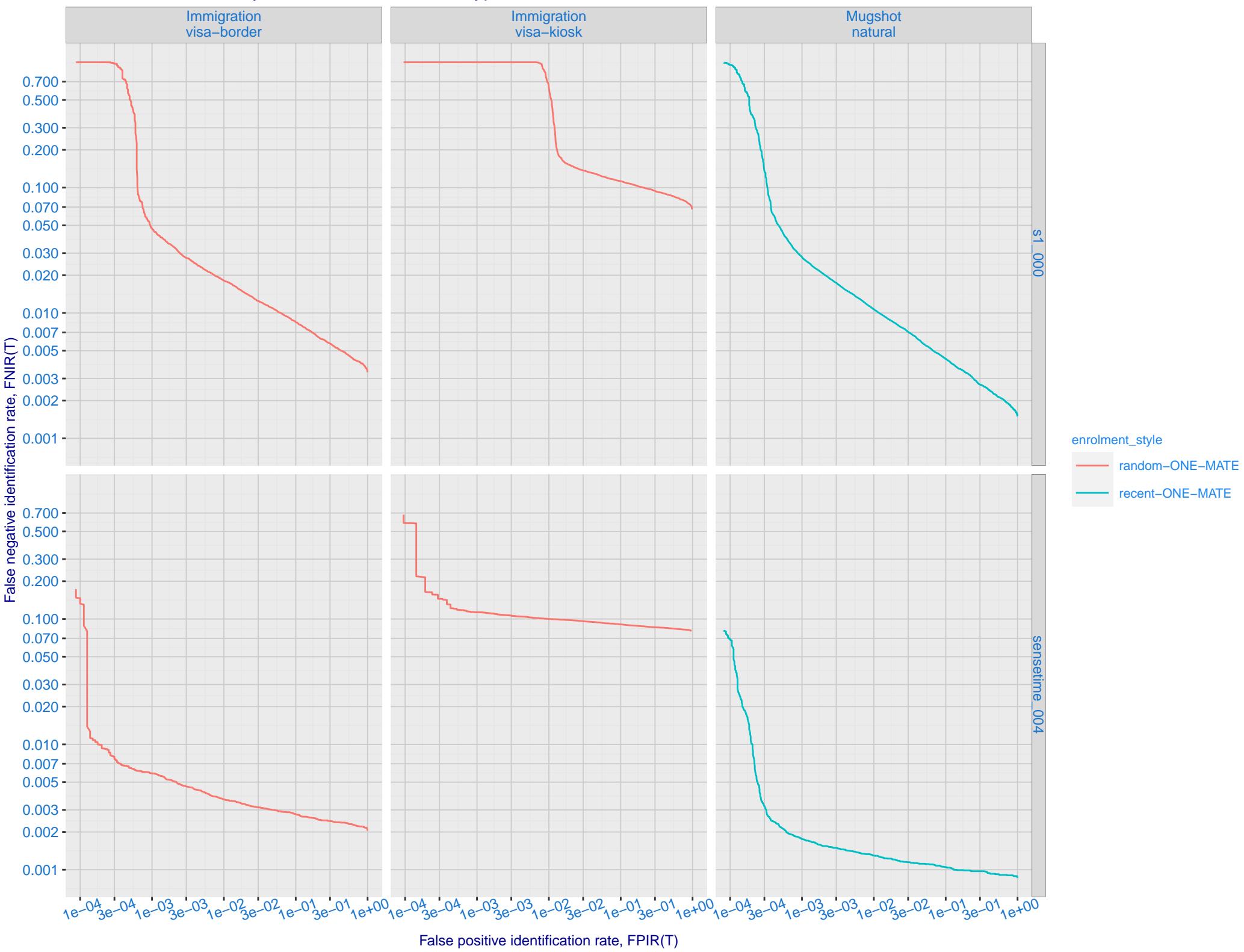
B: Mugshot natural images, identification mode: FNIR(N, L+1, T) vs. most accurate (sensetime\_004)



### C: Evolution of accuracy for S1 algorithms on three datasets 2018 – present

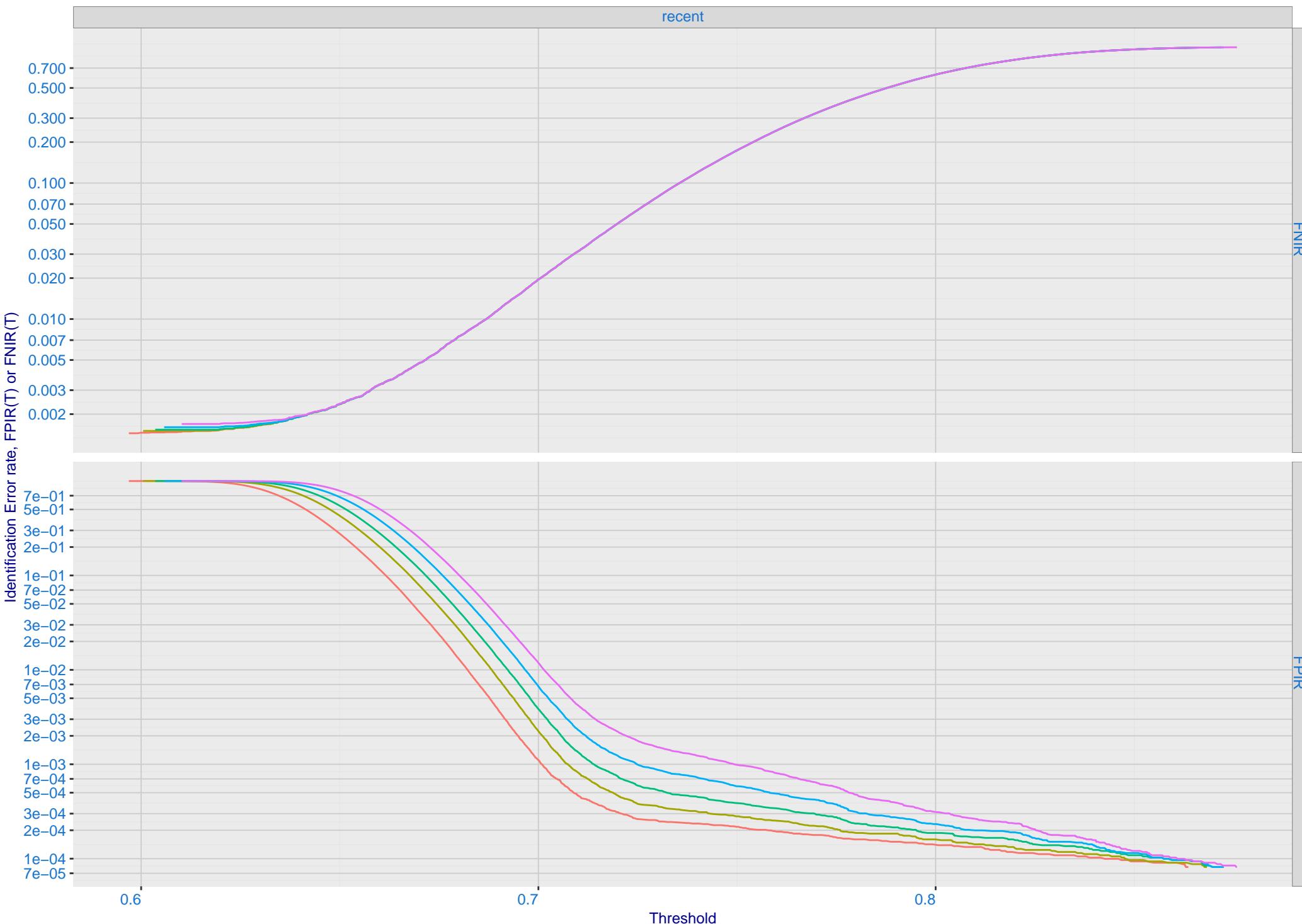


D: 1:N error tradeoff by dataset and enrollment type. N = 1600000 individuals

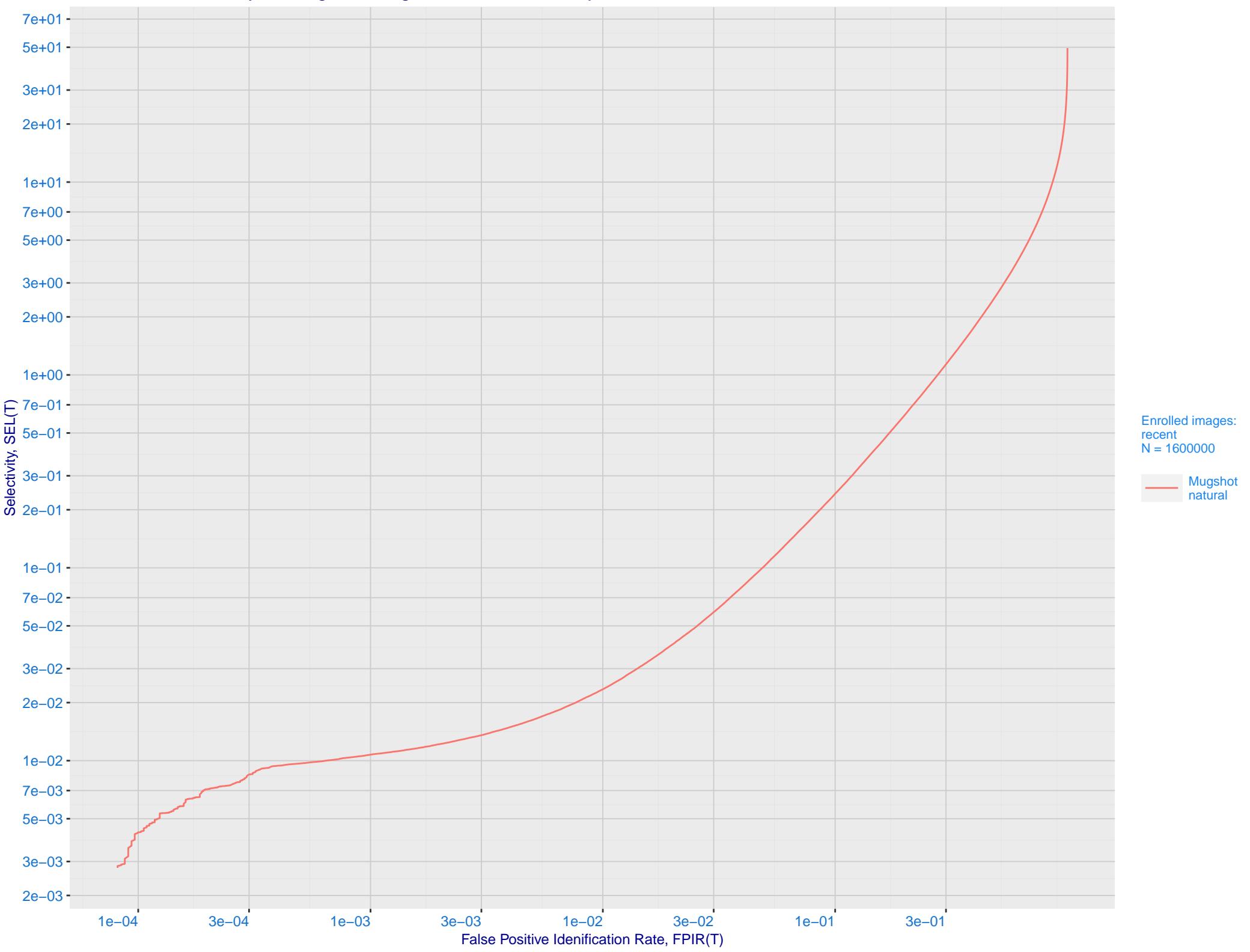


E: Dependence of error rates on T by number enrolled identities, N, for Mugshot natural images

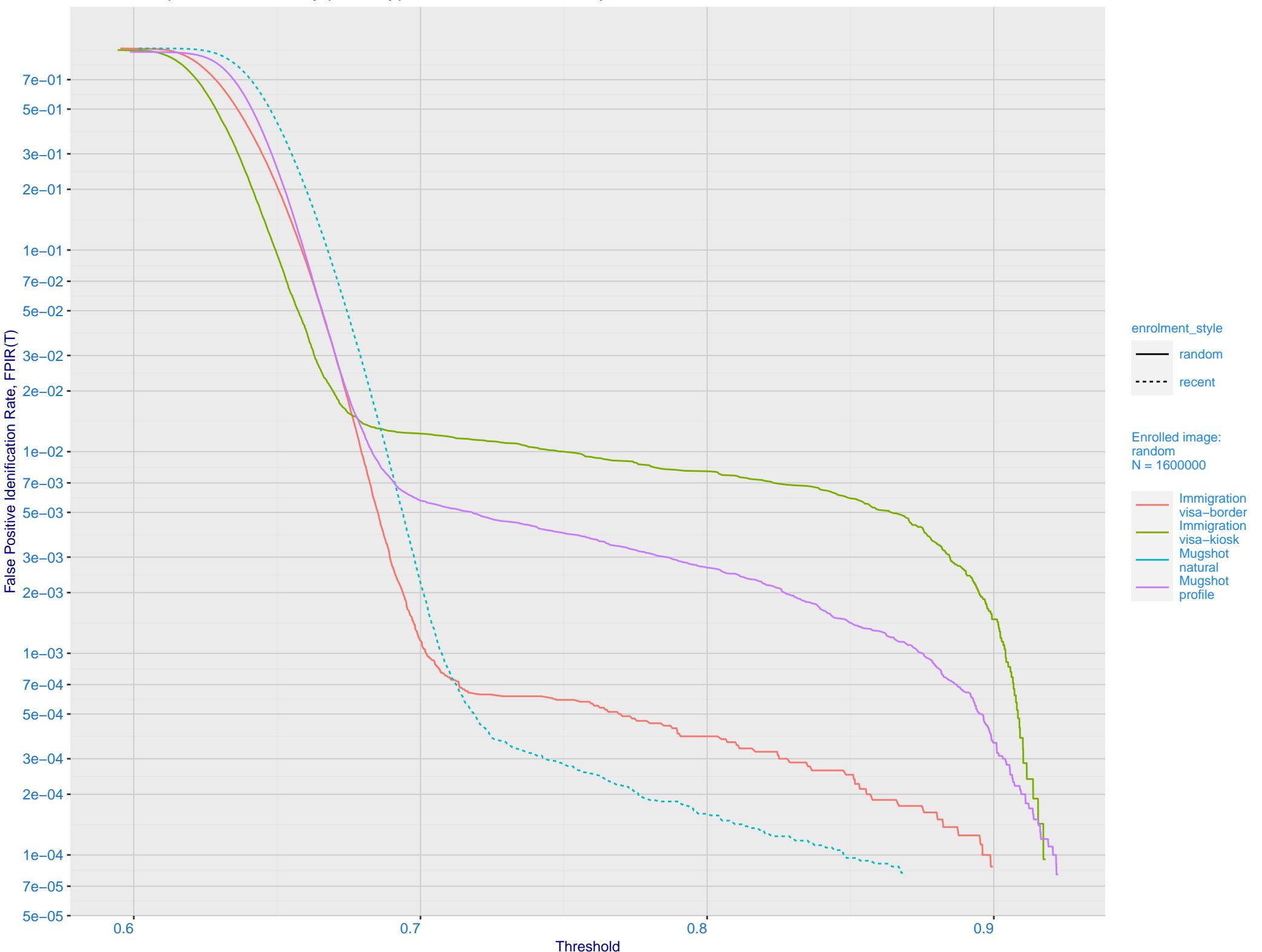
N 00640000 01600000 03000000 06000000 12000000



F: FPIR vs. Selectivity for mugshot images, N = 1600000 subjects enrolled with one recent mate

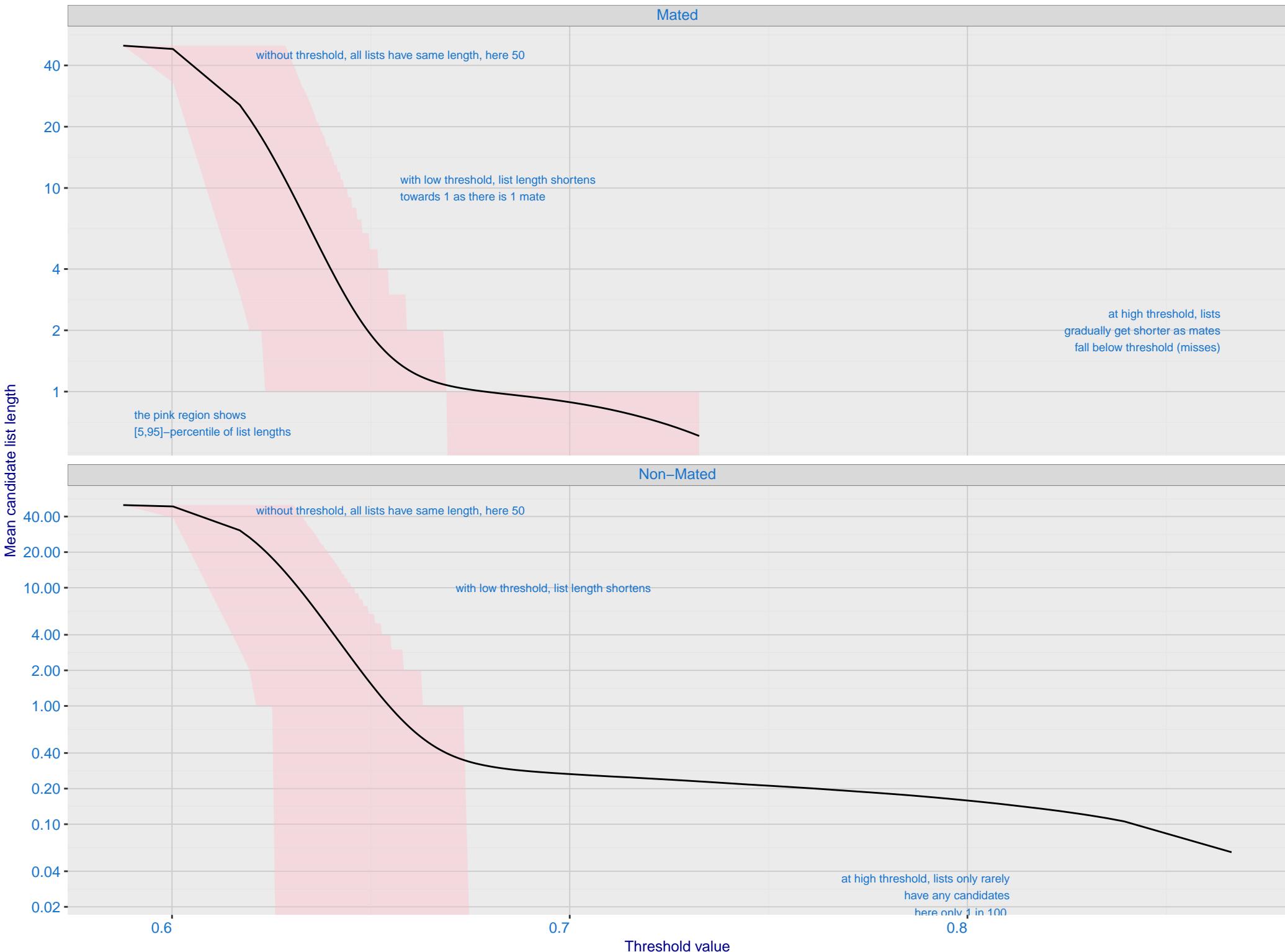


G: FPIR dependence on T by probe type for N = 1600000 subjects



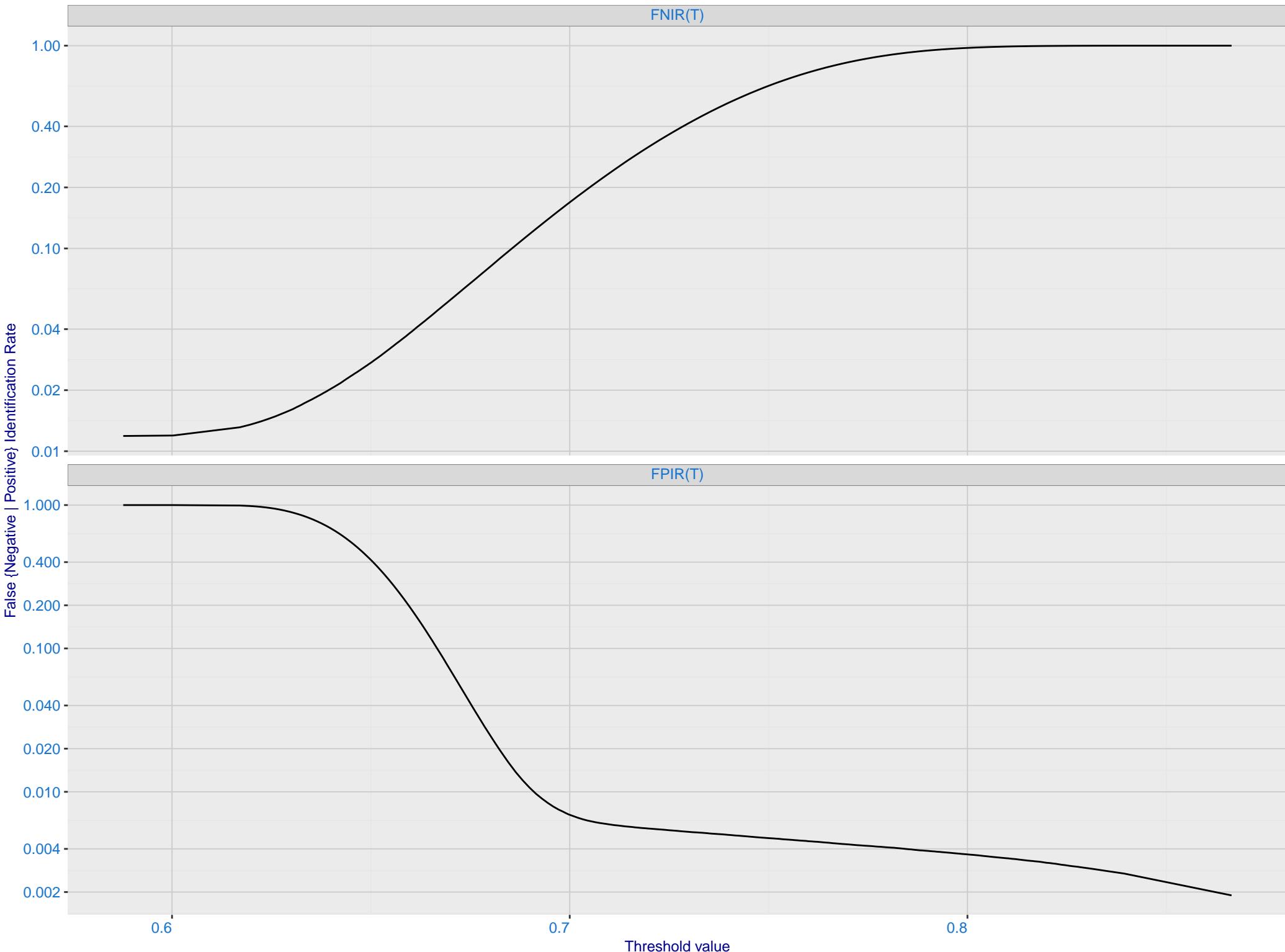
# H: Reduced length candidate lists for human review

Dataset is border–border with time–lapse [10,15] YRS with N = 1600000. Probes are 10–15 years later than enrollment image

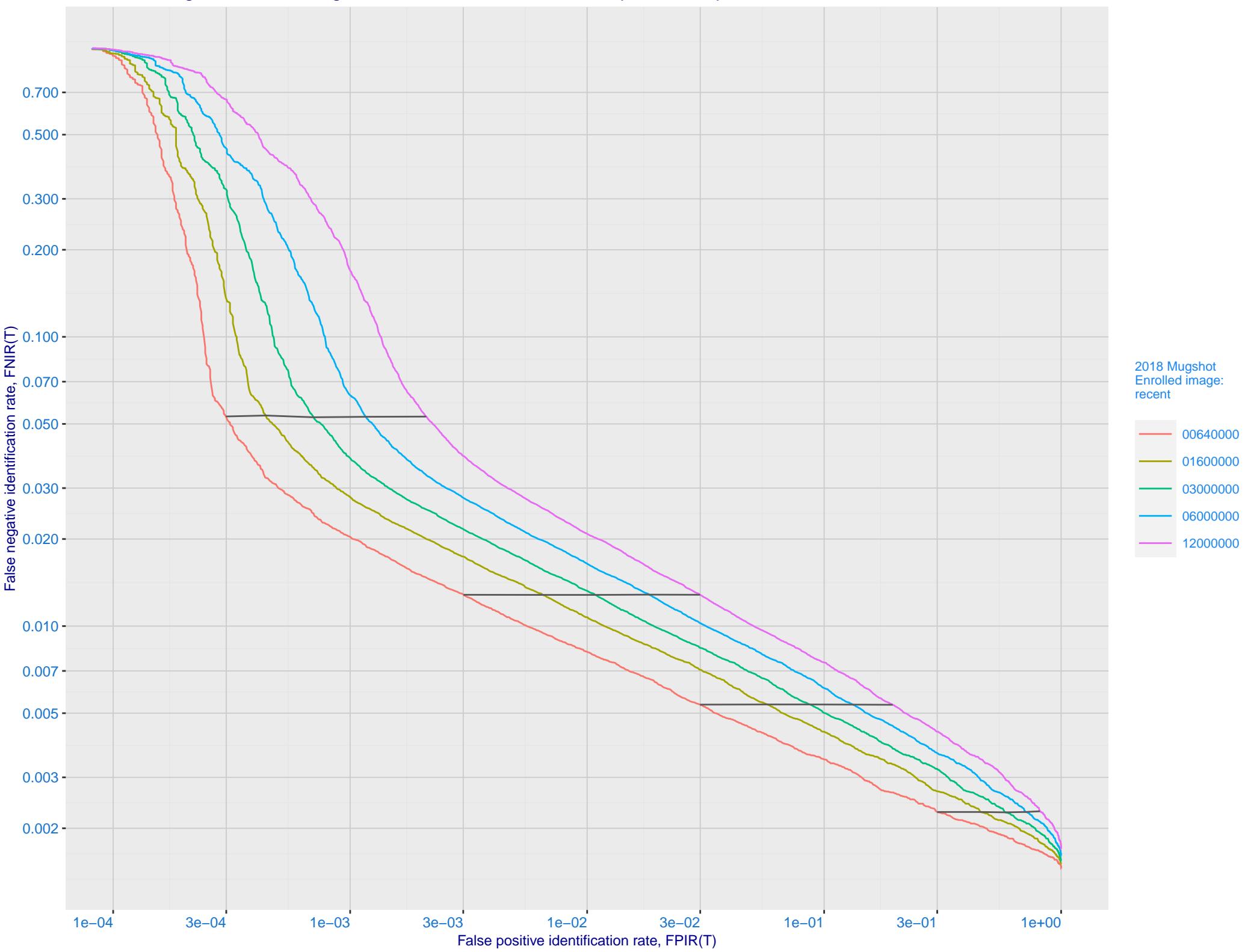


# I: FNIR and FPIR dependence on threshold

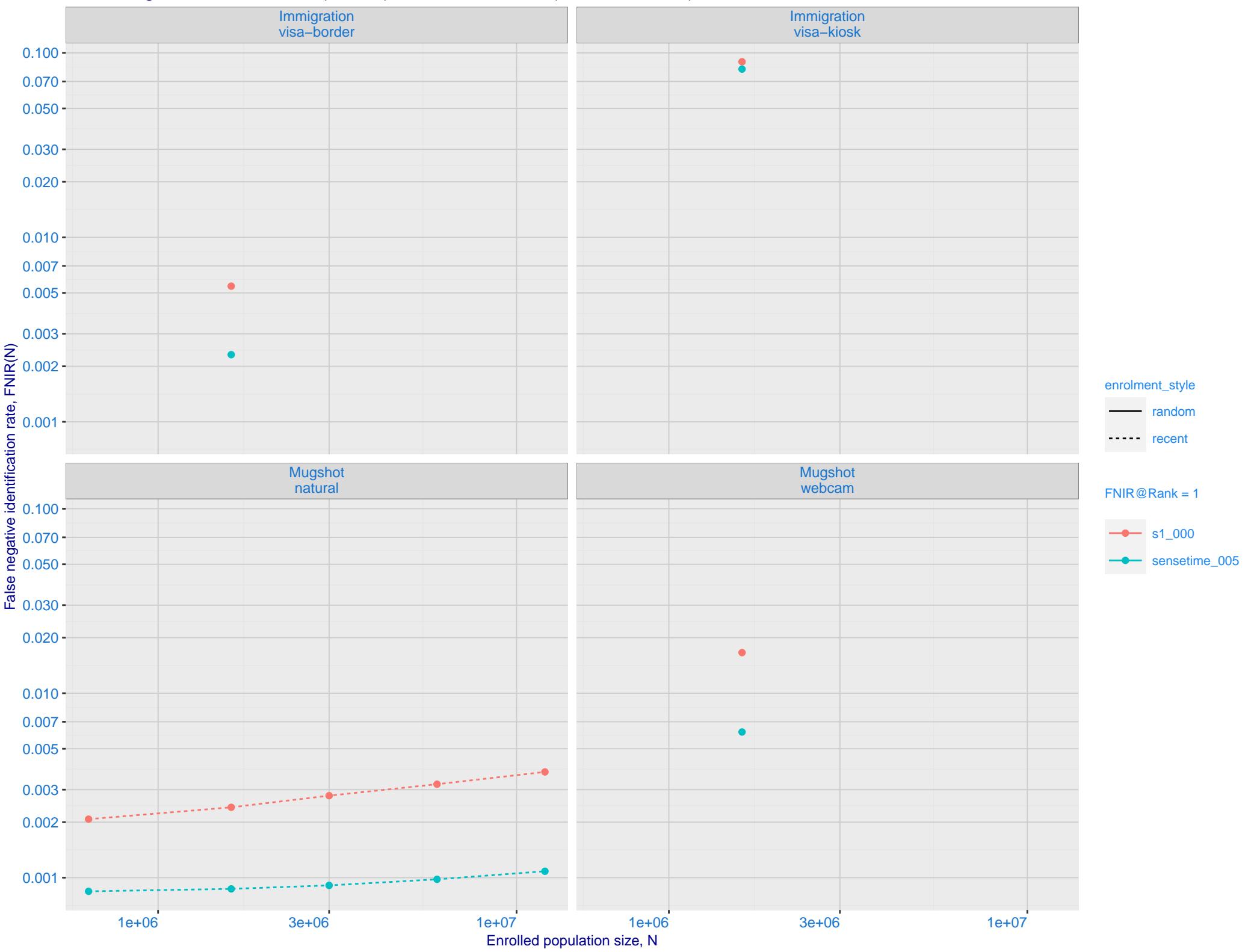
Dataset is border–border with time–lapse [10,15] YRS with N = 1600000. Probes are 10–15 years later than enrollment image



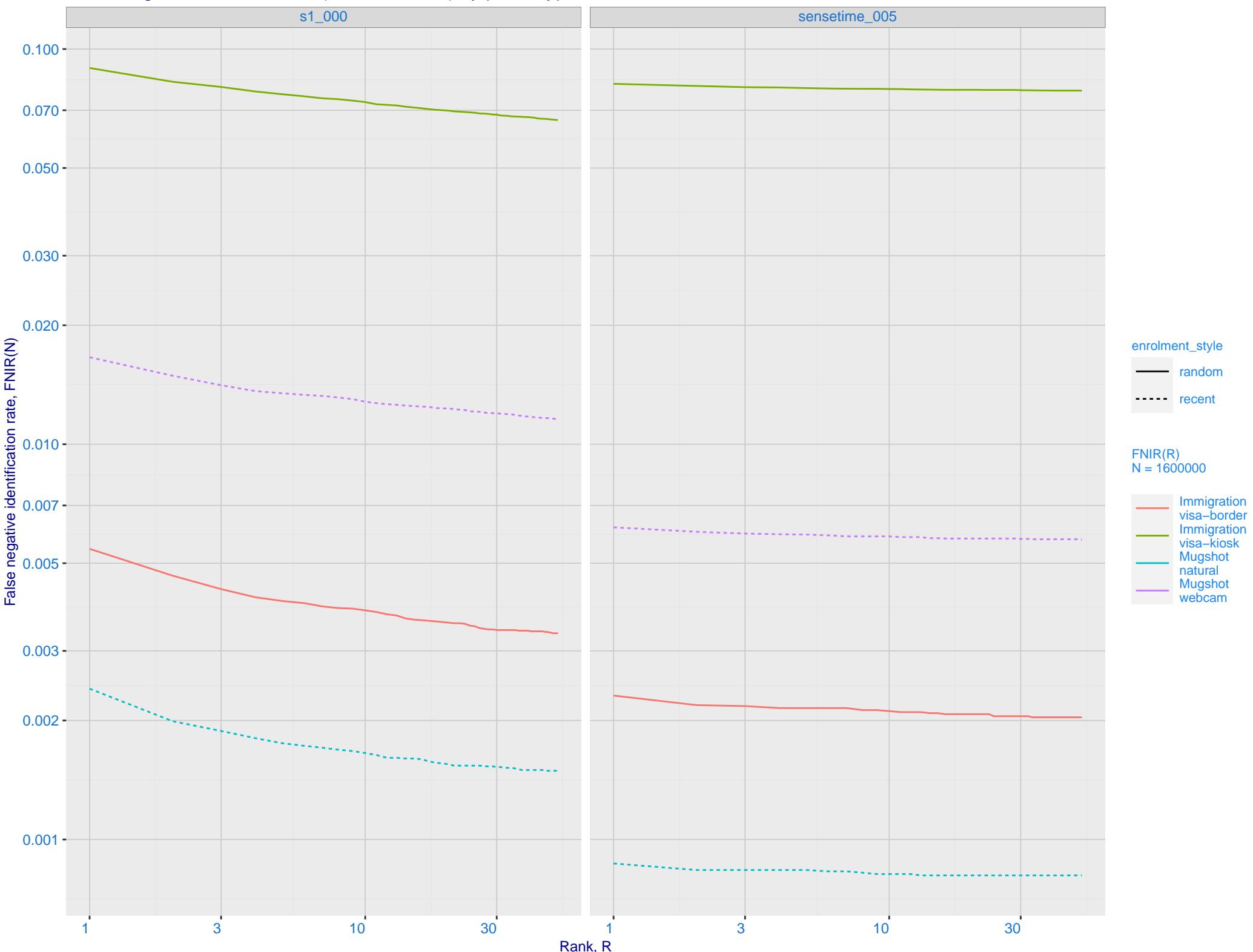
J: DET for Mugshot natural images and various N. Links connect points of equal threshold.



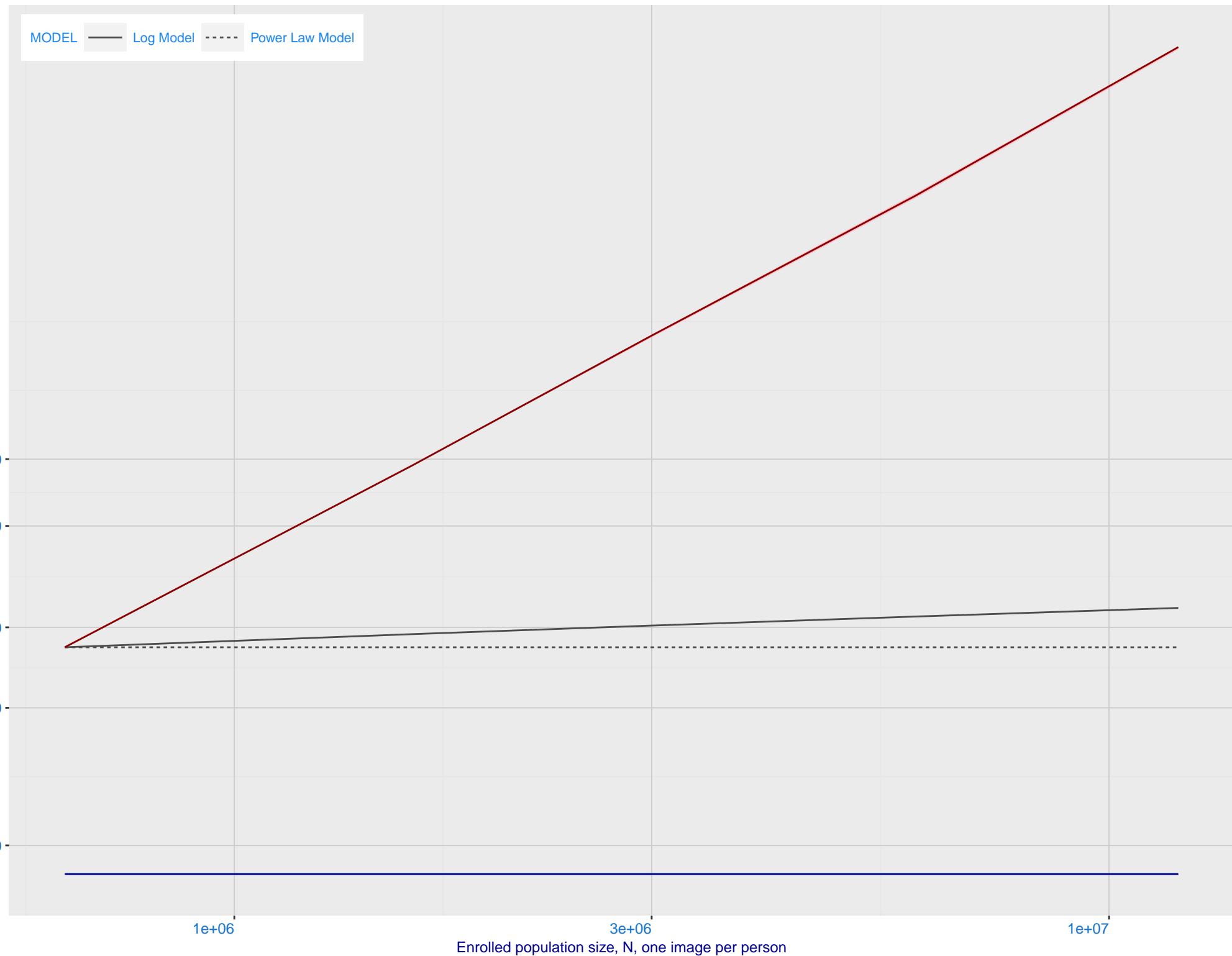
K: Investigational mode: FNIR(N, 1, 0) vs. most accurate (sensetime\_005)



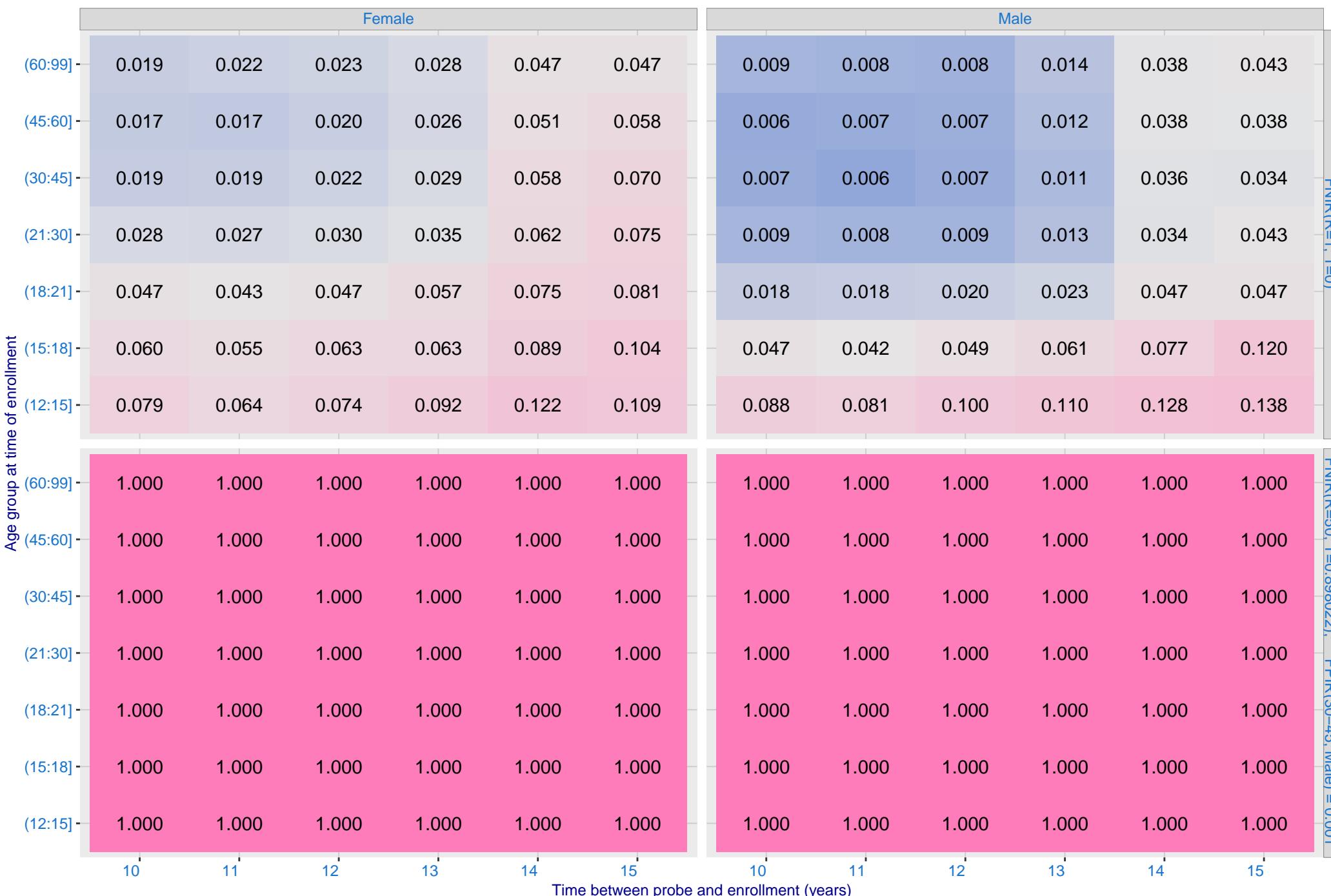
L: Investigational mode: FNIR(1600000, R, 0) by probe type



M: Template duration; search duration vs. N. The blue and pink ribbon covers 95 percent of observed measurements.  
The template generation time is independent of N. The log and power-law models are fit to the first two (N,T) observations



O: FNIR( $T$ ,  $N = 1.6$  million) by sex, age and time-lapse. The top row gives investigational rank-1 miss rates. The bottom panels give high threshold for more lights-out identification with low FPIR.



P: FPIR(N = 1.6 million) by sex and age. It is typical for false positive identification rates to be higher in women except in their teens.

Algorithm: s1\_000, Dataset: Border–Crossing Ageing  
Threshold: 0.898022 set to achieve FPIR(30–45, Male) = 0.001

Color encodes log(FPIR)



(The age of the highest non-mates will usually be similar to that of the probe.)

(60:99] 0.0008 0.0008

(45:60] 0.0010 0.0011

(30:45] 0.0011 0.0010

(21:30] 0.0014 0.0008

(18:21] 0.0005 0.0005

(15:18] 0.0003 0.0002

(12:15] 0.0001 0.0001

Female

Male

Sex of person in non-mate probe  
(The sex of the highest non-mates will usually be that of the probe.)

Q: Identification FNIR(N, T, L+1) and Investigational FNIR(N, 0, R) under ageing

Dataset: 2018 Mugshot N = 3068801

